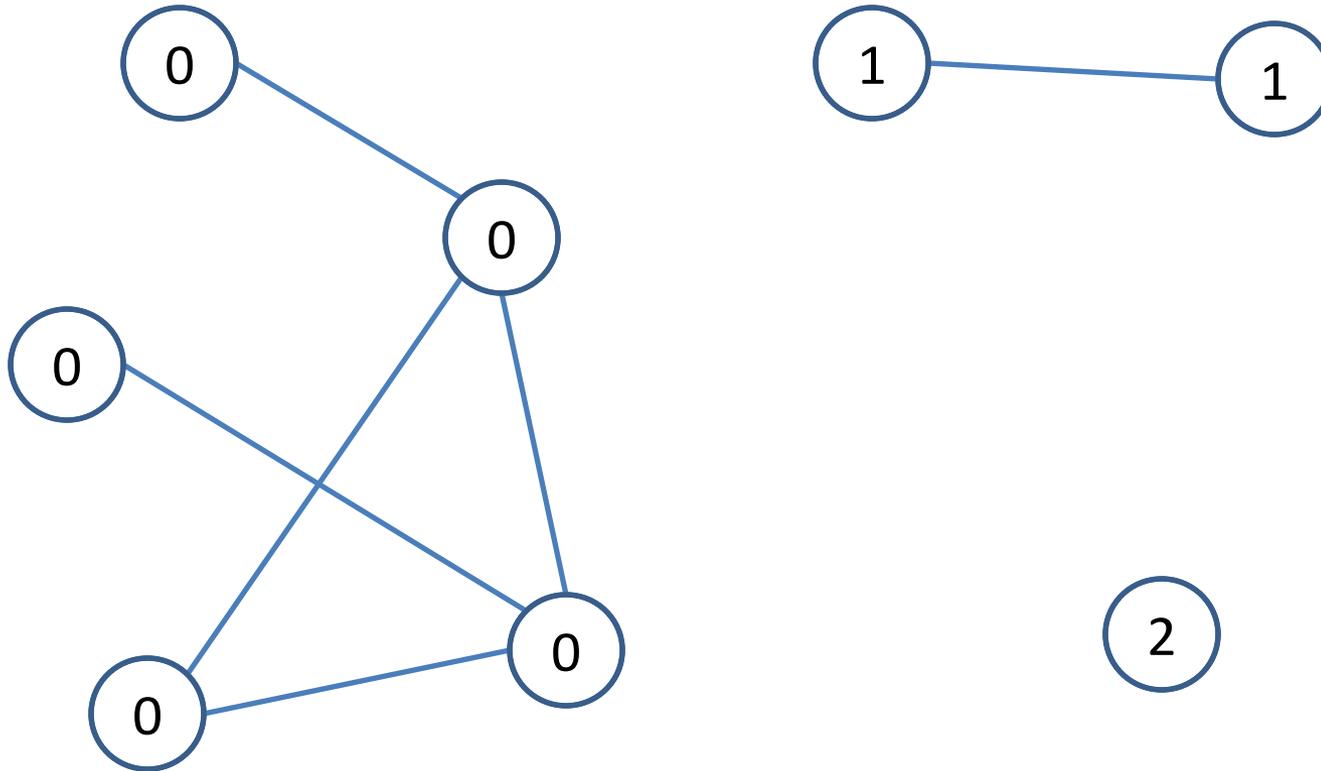


# A Simple and Practical Linear-Work Parallel Algorithm for Connectivity

Julian Shun, Laxman Dhulipala, and Guy Blelloch

# Connected Component Labeling



Given an undirected graph, label all vertices such that  $L(u) = L(v)$  if and only if there is a path between  $u$  and  $v$

# Connected Component Labeling

- What are some simple algorithms?
  - Depth-first search
    - Linear work/span
    - Versions of DFS that are parallel are not work-efficient
  - Breadth-first search
    - Linear work
    - Parallelism limited by graph diameter
    - Polylogarithmic span version not work-efficient
  - Spanning forest
    - Good parallelism
    - Practical parallel implementations not linear work

# Connected Component Labeling

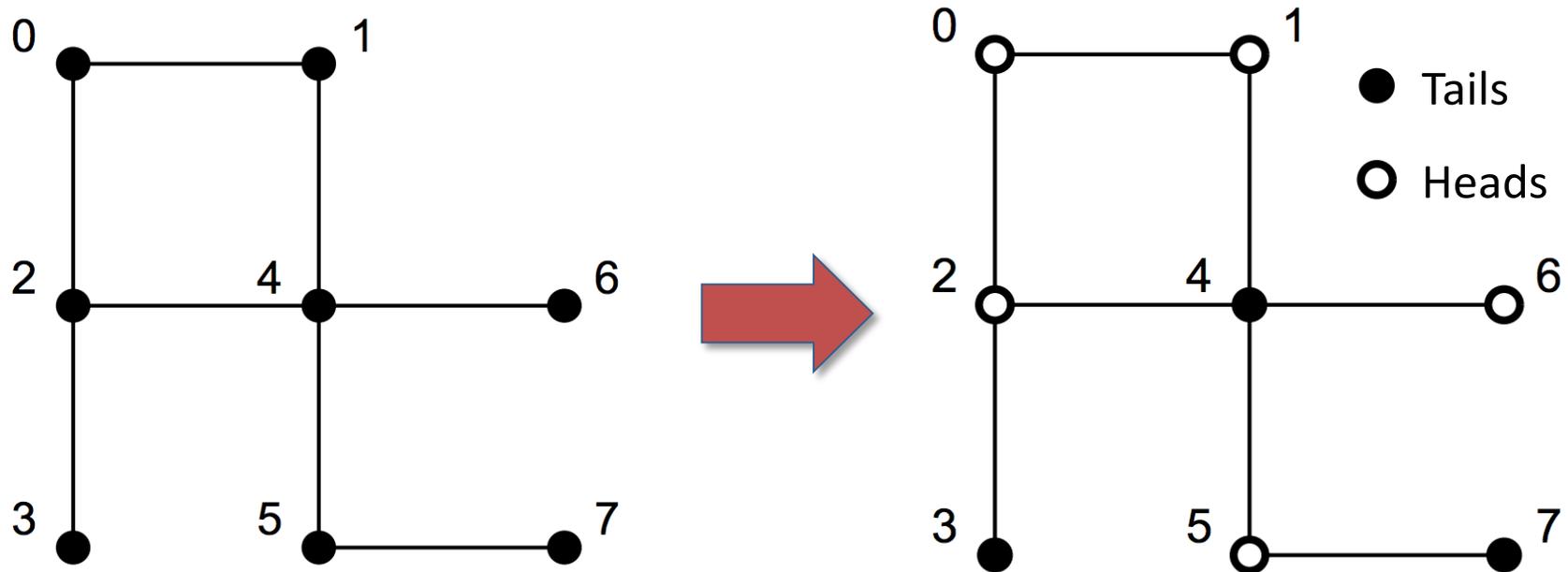
- Parallel (polylogarithmic span) algorithms
  - Shiloach and Vishkin, Awerbuck and Shiloach
    - Combines (contracts) vertices in each iteration
    - $O(m \log n)$  work,  $O(\log n)$  span
  - Reif, Phillips
    - Uses randomization to simplify contraction algorithms
    - $O(m \log n)$  work and  $O(\log n)$  span w.h.p.
    - $O(\log n)$  rounds but don't guarantee a constant fraction of edges removed
  - $O(m)$  work algorithms
    - Gazit '91, Halperin/Zwick '96, Cole et al. '96, Poon/Ramachandran '97, Pettie/Ramachandran '02
    - Quite complicated. No one has implemented these

# Our Contributions

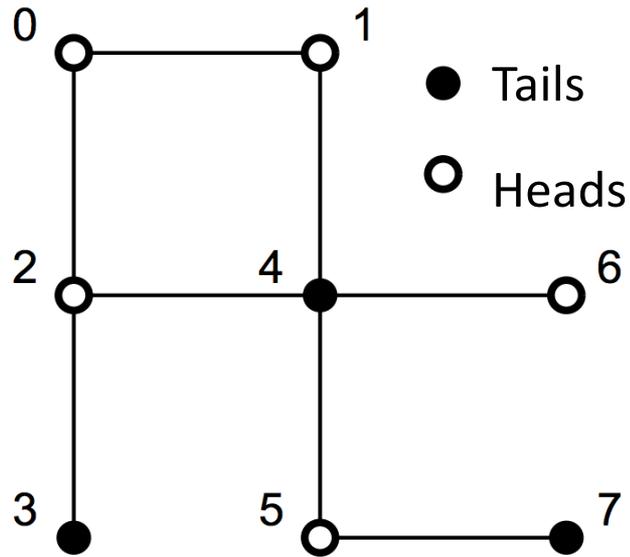
- Practical parallel connectivity algorithm with linear work and polylogarithmic span
- Experimental evaluation: competitive with existing parallel implementations (that are not linear-work and polylogarithmic span)

# Previous Work: Random Mate

- Idea: Form a set of non-overlapping star subgraphs and contract them
- Each vertex flips a coin. For each Heads vertex, pick an arbitrary Tails neighbor (if there is one) and point to it

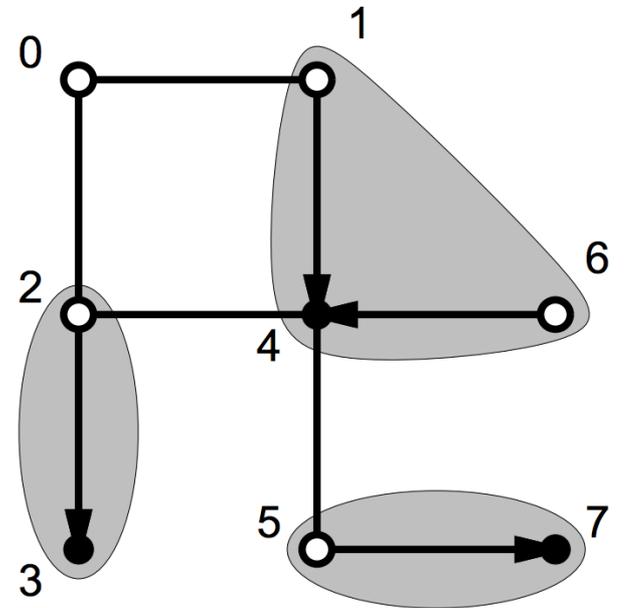
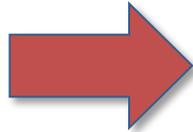


# Previous Work : Random Mate

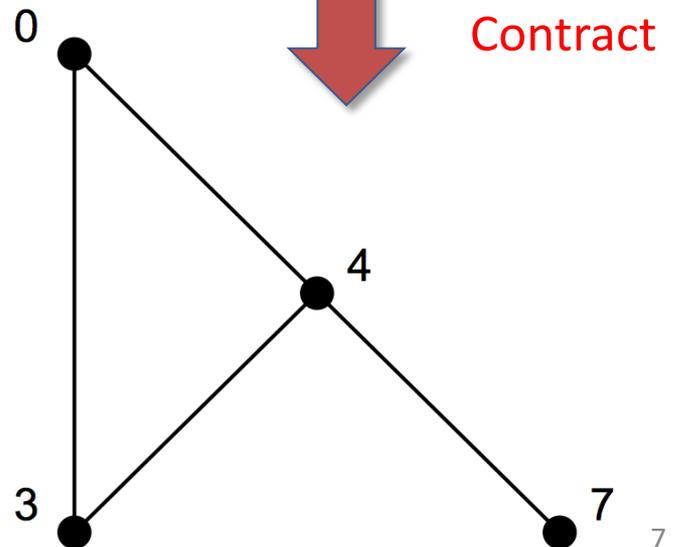


● Tails  
○ Heads

Form stars



Contract



Repeat until each component has a single vertex

Expand vertices back in reverse order with label of neighbor

# Previous Work : Random Mate

CC\_Random\_Mate(L, E)

if( |E| = 0) Return L //base case

else

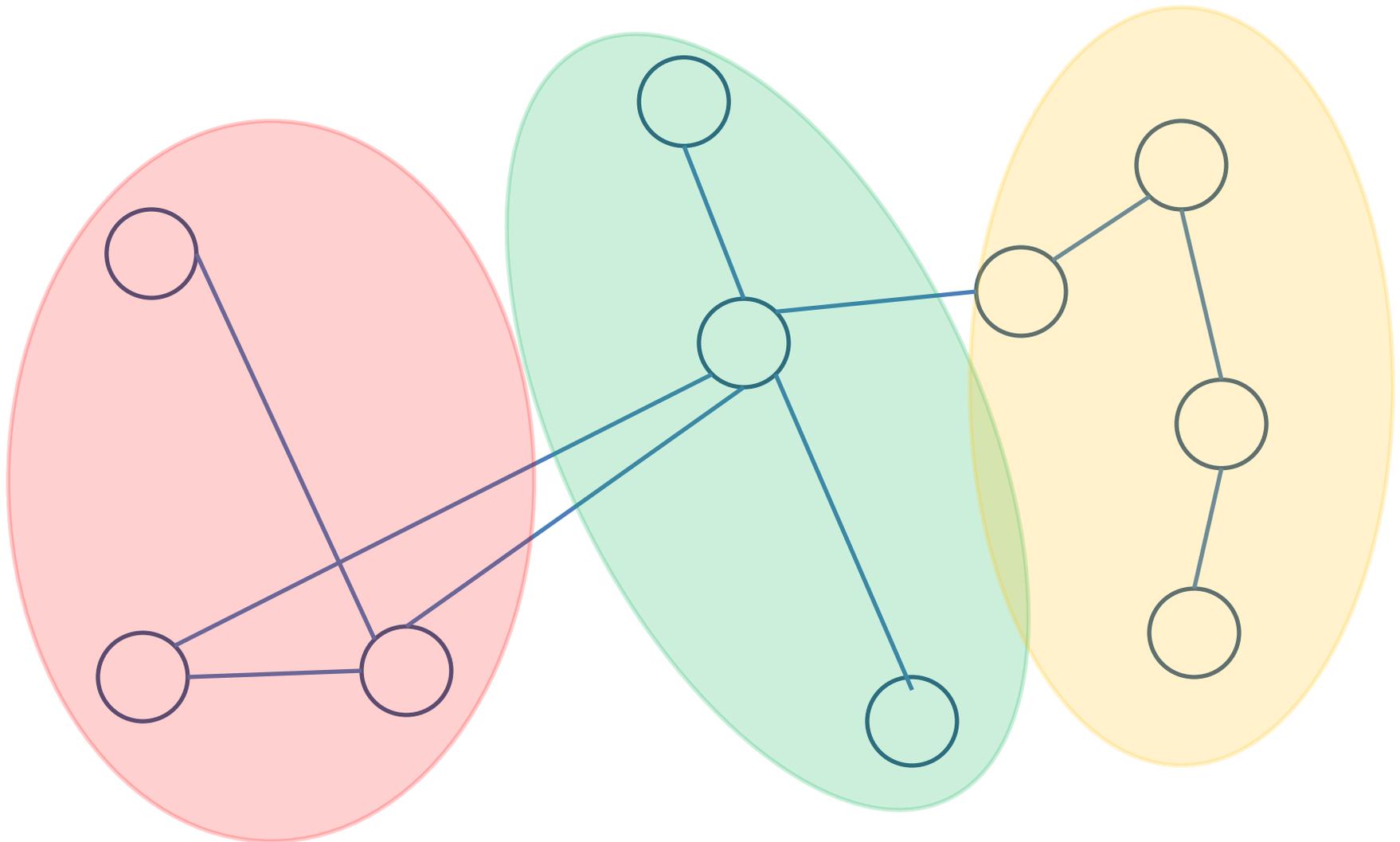
1. Flip coins for all vertices
2. For  $v$  where  $\text{coin}(v)=\text{Heads}$ , hook to arbitrary Tails neighbor  $w$  and set  $L(v) = L(w)$
3.  $E' = \{ (L(u), L(v)) \mid (u, v) \in E \text{ and } L(u) \neq L(v) \}$
4.  $L' = \text{CC\_Random\_Mate}(L, E')$
5. For  $v$  where  $\text{coin}(v)=\text{Heads}$ , set  $L'(v) = L'(w)$  where  $w$  is the Tails neighbor that  $v$  hooked to in Step 2
6. Return  $L'$

- Each iteration requires  $O(m+n)$  work and  $O(1)$  span
  - Assumes we do not pack vertices and edges
- Each iteration eliminates at least  $1/4$  of the vertices in expectation  $\rightarrow O(\log n)$  rounds w.h.p.

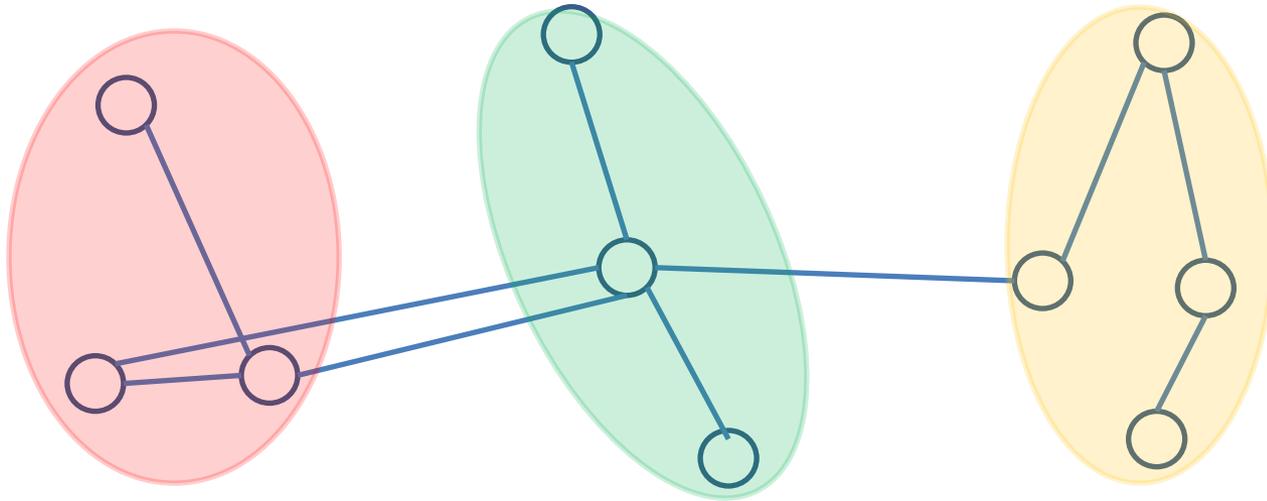
$$W = O(m \log n) \text{ w.h.p.}$$

$$S = O(\log n) \text{ w.h.p.}$$

# Low diameter decomposition



# Low diameter decomposition

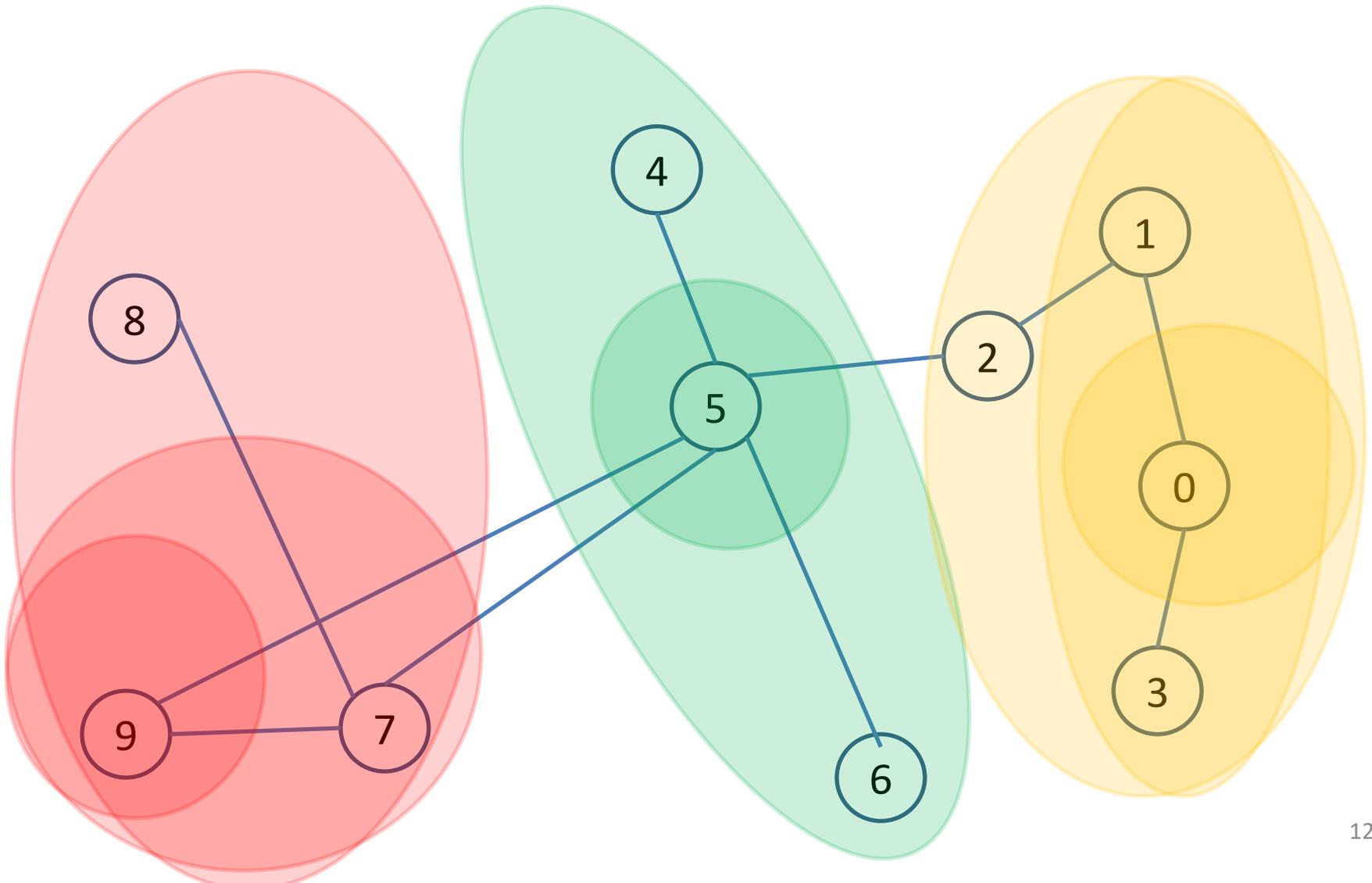


- $(\beta, d)$ -decomposition ( $0 < \beta < 1$ ) partitions  $V$  into  $V_1, \dots, V_k$  such that
  - The shortest path between any two vertices in a partition is at most  $d$
  - The number of inter-partition edges is at most  $\beta m$
- Used in linear system solvers and metric embeddings

# Low diameter decomposition

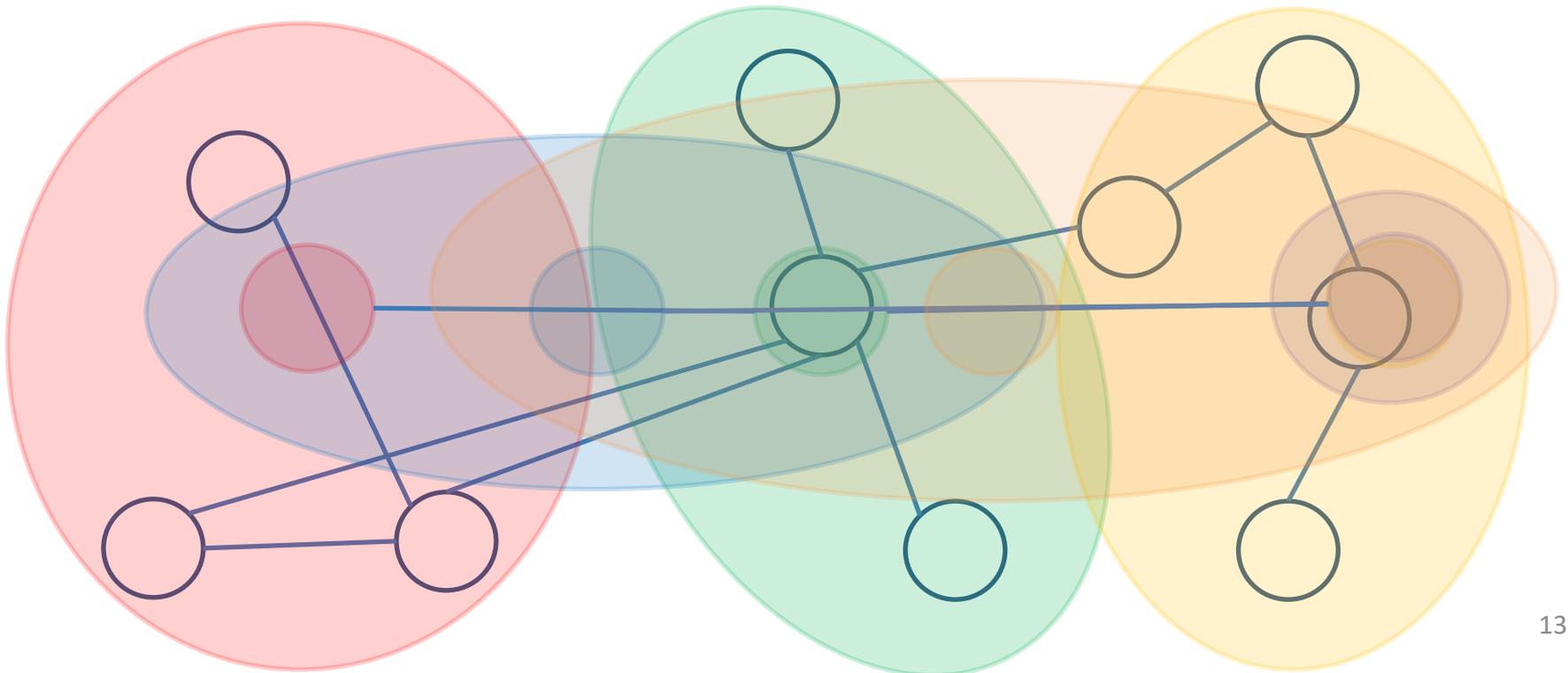
- A  $(\beta, O(\log n / \beta))$ -decomposition can be computed in  $O(m)$  expected work and  $O(\log^2 n / \beta)$  span w.h.p. [Miller et al. 2013]
  - Start breadth-first searches from vertices with exponentially-distributed (parameter  $\beta$ ) start times
    - Each BFS creates a partition containing the source and all vertices explored
    - A BFS does not explore vertices already visited by another BFS
    - All vertices will have started BFS or been explored by time  $O(\log n / \beta)$
  - BFS's are work-efficient and terminate in  $O(\log n / \beta)$  iterations.
    - Each iteration requires  $O(\log n)$  span.
  - Bounding number of inter-partition edges:
    - An edge is inter-partition if the first two BFS's that reach it do so within a one time step of each other
    - Probability that this happens is at most  $\beta$  due to properties of exponential distribution
    - Linearity of expectations gives at most  $\beta m$  edges cut

# Low diameter decomposition example



# Our Connectivity Algorithm

- Compute a  $(\beta, O(\log n / \beta))$ -decomposition
- Contract each partition into a single vertex
- Recurse



# Our Connectivity Algorithm

- Compute a  $(\beta, O(\log n / \beta))$ -decomposition
- Contract each partition into a single vertex
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## Analysis for $\beta=1/2$

- Assume contraction can be done in linear work and in  $O(\log n)$  span
- $m/2$  edges remain after each round in expectation
  - Work =  $O(m) + O(m/2) + \dots = O(m)$  in expectation
- $O(\log n)$  levels of recursion suffice w.h.p.
  - Span =  $O(\log n) * O(\log^2 n / \beta) = O(\log^3 n)$  w.h.p.

# Contraction

- Contraction can be done in  $O(\log n)$  span with bookkeeping and parallel prefix sums
  - Intra-partition edges are filtered out in  $O(m)$  work and  $O(\log n)$  span
  - Prefix sums: relabel vertices to smaller range
  - Duplicate edges removed using parallel hashing in  $O(m)$  work and  $O(\log n)$  span
    - Not needed theoretically

# Improving span

- Each round of BFS can be implemented in  $O(\log^* n)$  span w.h.p. using approximate prefix sum and compaction [Gil-Matias-Vishkin '91, Goodrich-Matias-Vishkin '94]
  - Improves span of low diameter decomposition to  $O(\log n \log^* n)$
- Recurse for  $O(\log \log n)$  rounds
  - Left with  $O(m/\log n)$  edges
  - Switch to  $O(m \log n)$  work,  $O(\log n)$  span algorithm
- Result: Linear work algorithm with  $O(\log n \log \log n \log^* n)$  span w.h.p.

# Low diameter decomposition variants

- Resolving conflicts among BFS's
  - Decomp-min: breaks ties deterministically
    - Miller et al. showed this produces  $(\beta, O(\log n/\beta))$ -decomposition
    - Uses write-with-min (via compare-and-swap)
    - Requires two phases
  - Decomp-arb: breaks ties arbitrarily
    - We prove  $(2\beta, O(\log n/\beta))$ -decomposition
    - Uses compare-and-swap
    - Requires just a single phase
  - Decomp-arb-hybrid: uses direction-optimizing BFS
    - This is the fastest one and used in the following experimental results

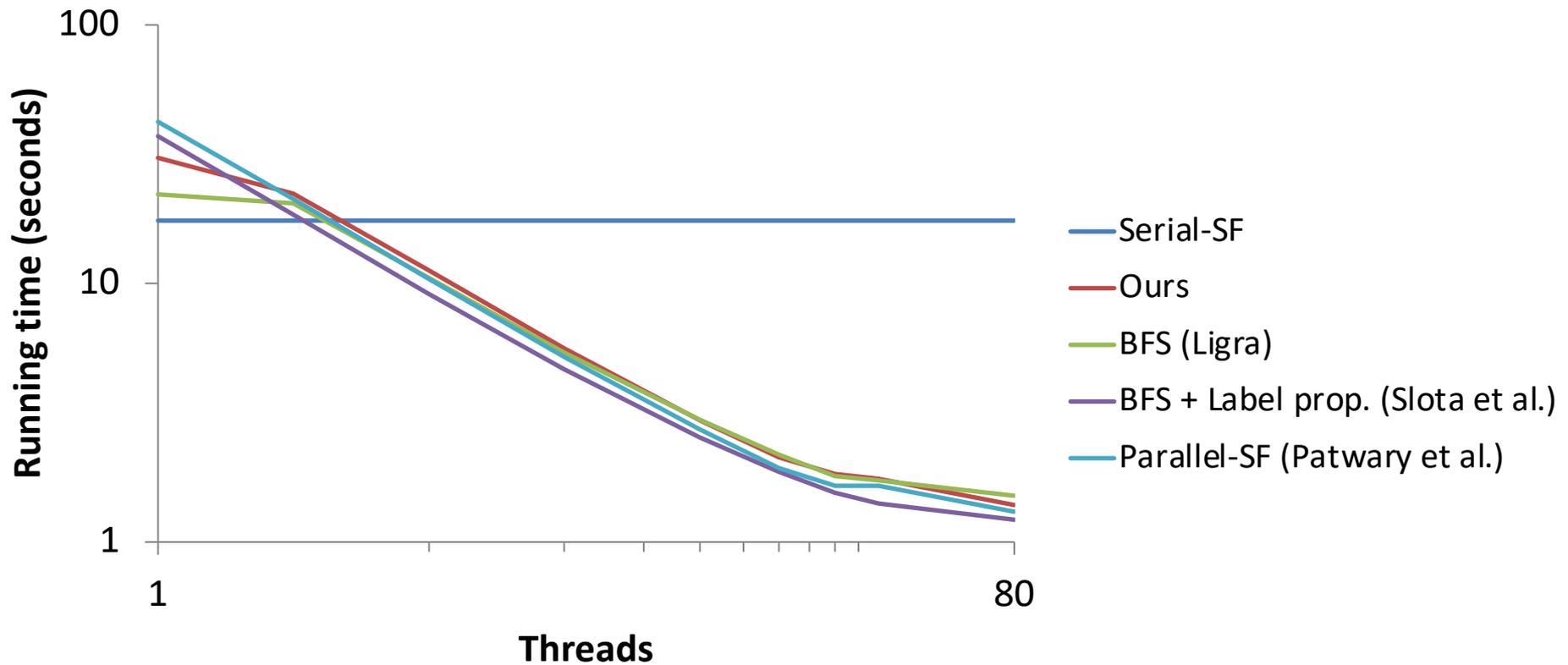
# Experiments

- 40-core (with 2-way hyper-threading) Intel Nehalem machine
- Implemented in Cilk Plus
- 3 different implementations, but only showing best one
- Real-world and artificial graphs

# Compare to existing implementations

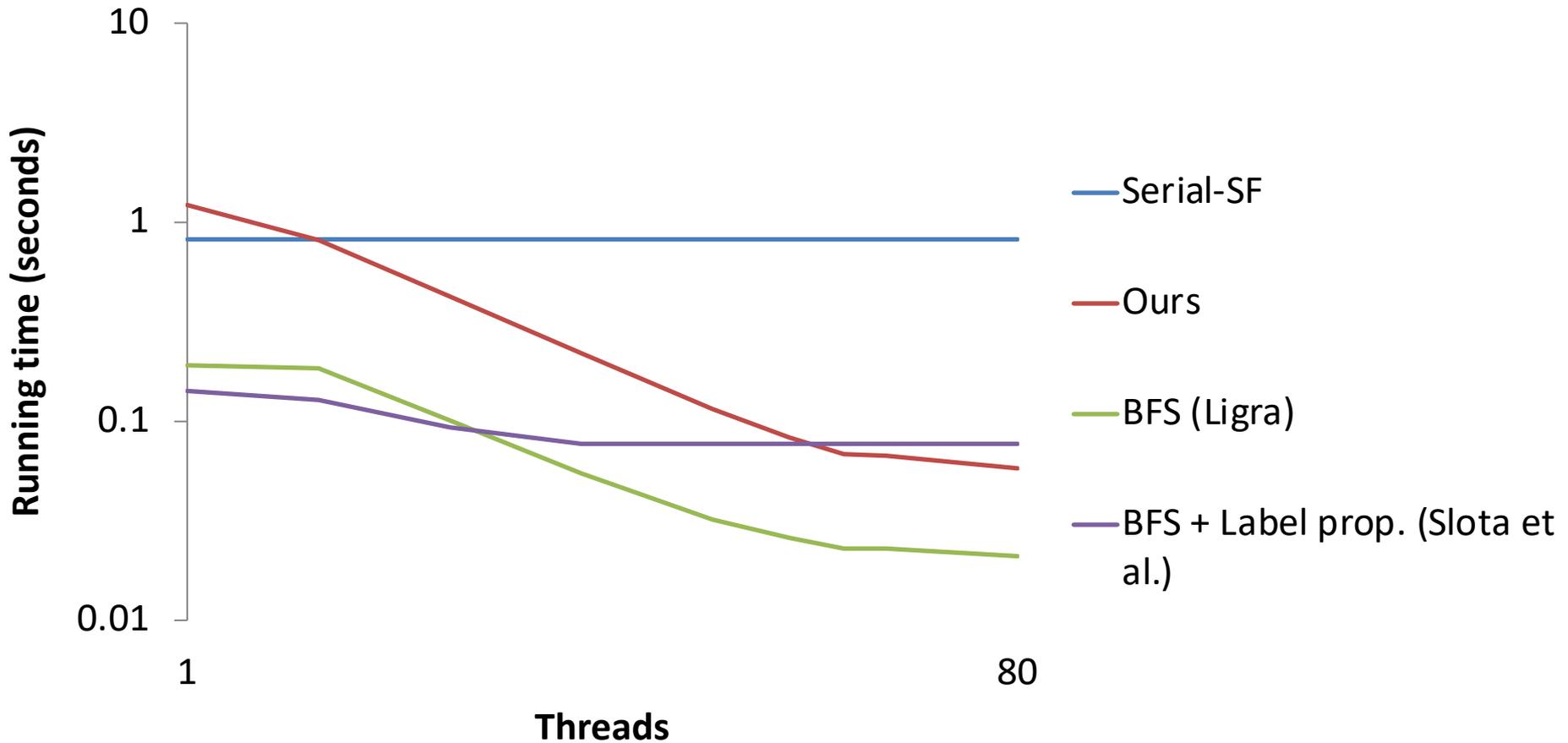
- Existing implementations
  - Sequential spanning forest
  - Parallel spanning forest (Problem Based Benchmark Suite)
  - Parallel spanning forest (Patwary et al.)
  - Parallel BFS (Ligra)
  - Parallel BFS + Label propagation (Slota et al.)
- None provably linear work and polylog span

# 3D grid graph ( $n = 10^8$ , $m = 3 \times 10^8$ )



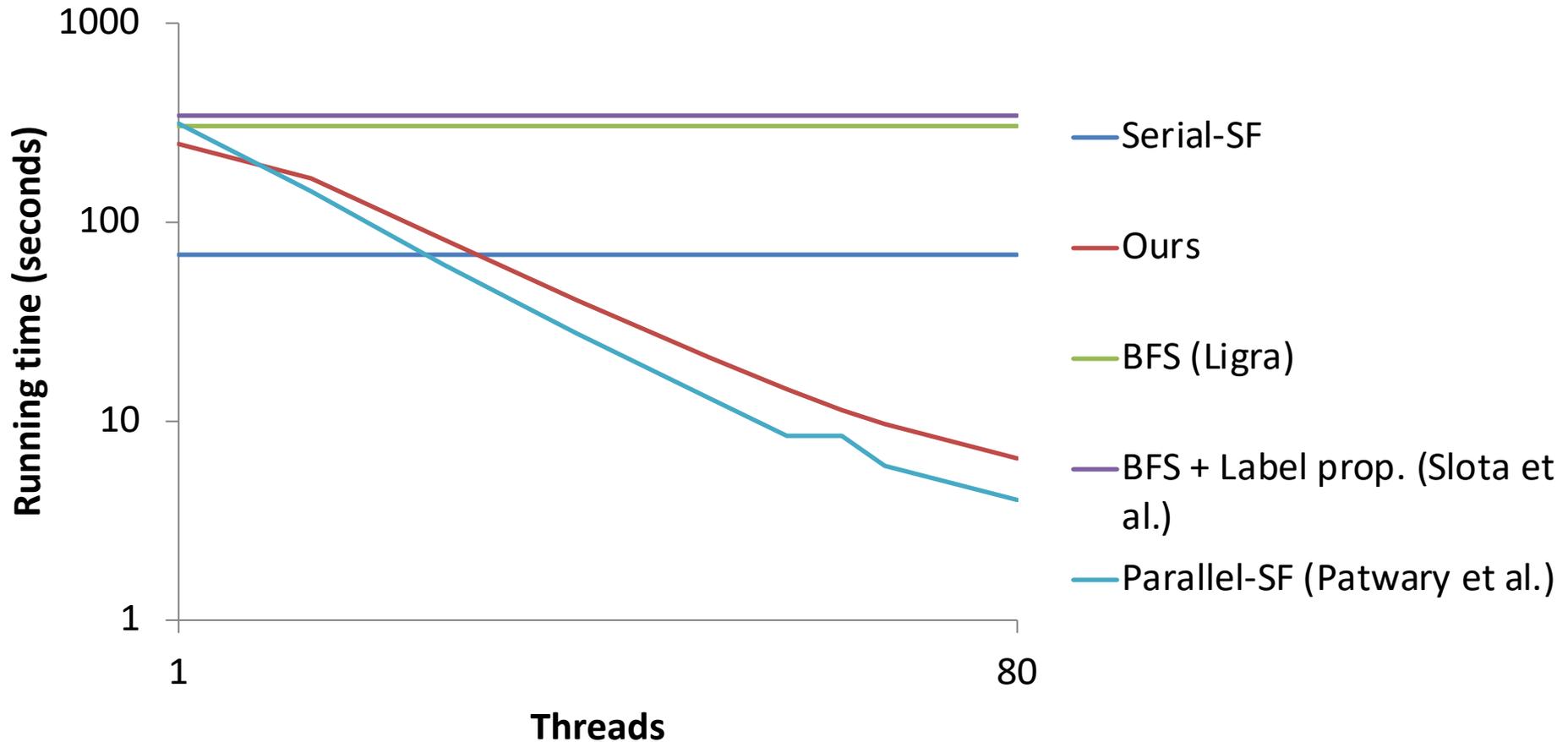
- Competitive with other implementations

# com-Orkut graph ( $n \approx 3 \times 10^6$ , $m \approx 10^8$ )



- Fastest implementation uses single BFS

# Line graph ( $n = 5 \times 10^8$ , $m = 5 \times 10^8$ )



- Algorithms based on single BFS do poorly

# Our algorithm is competitive

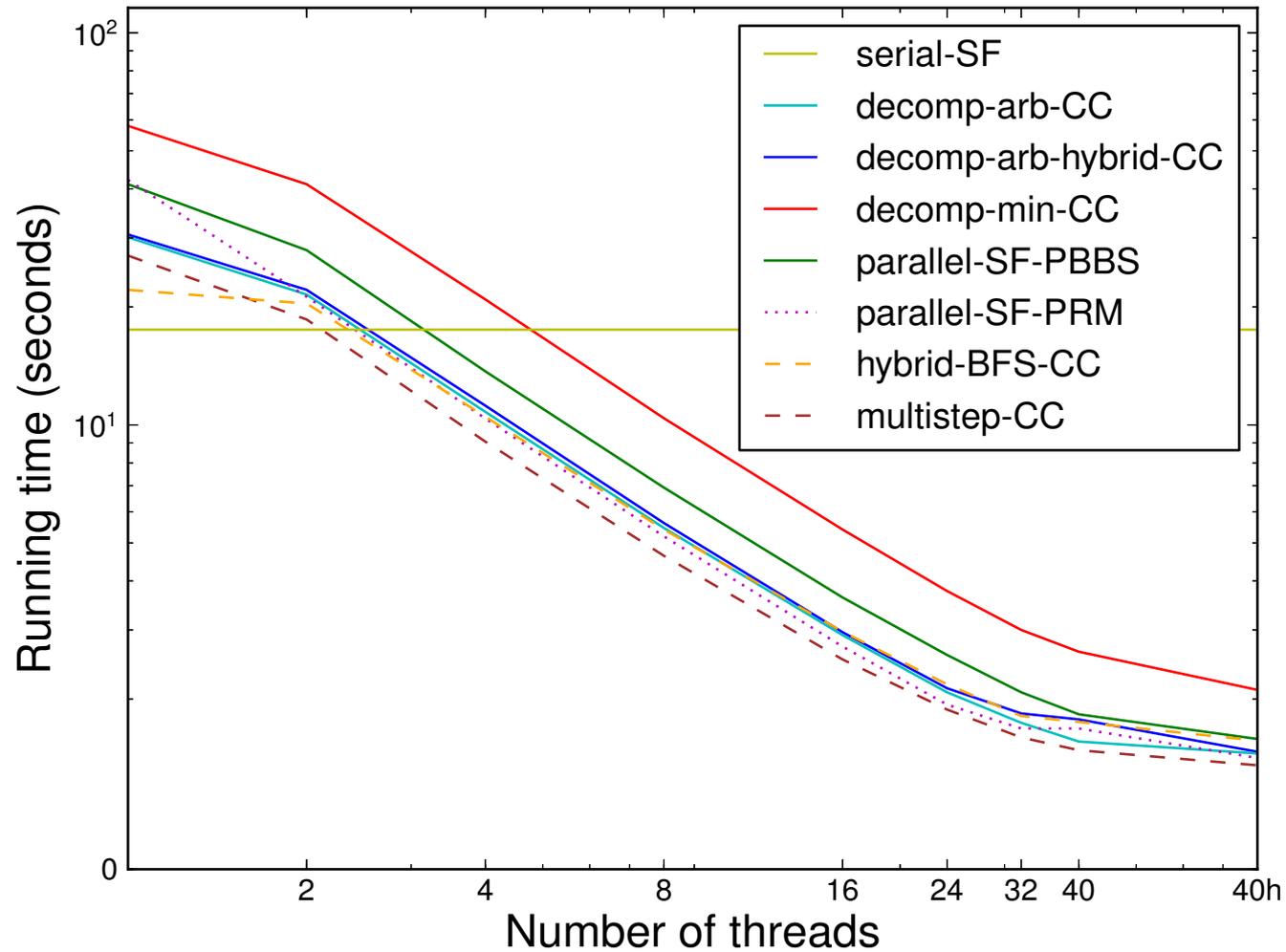
- No “worst-case” inputs
- Performance always close to the fastest implementation for any graph
  - Only at most 70% slower than spanning forest algorithms, and usually much less
  - Can be faster or slower than BFS, depending on graph diameter
- Up to 13x speedup on 40 cores relative to sequential
- 18—39x self-relative speedup

# Conclusion

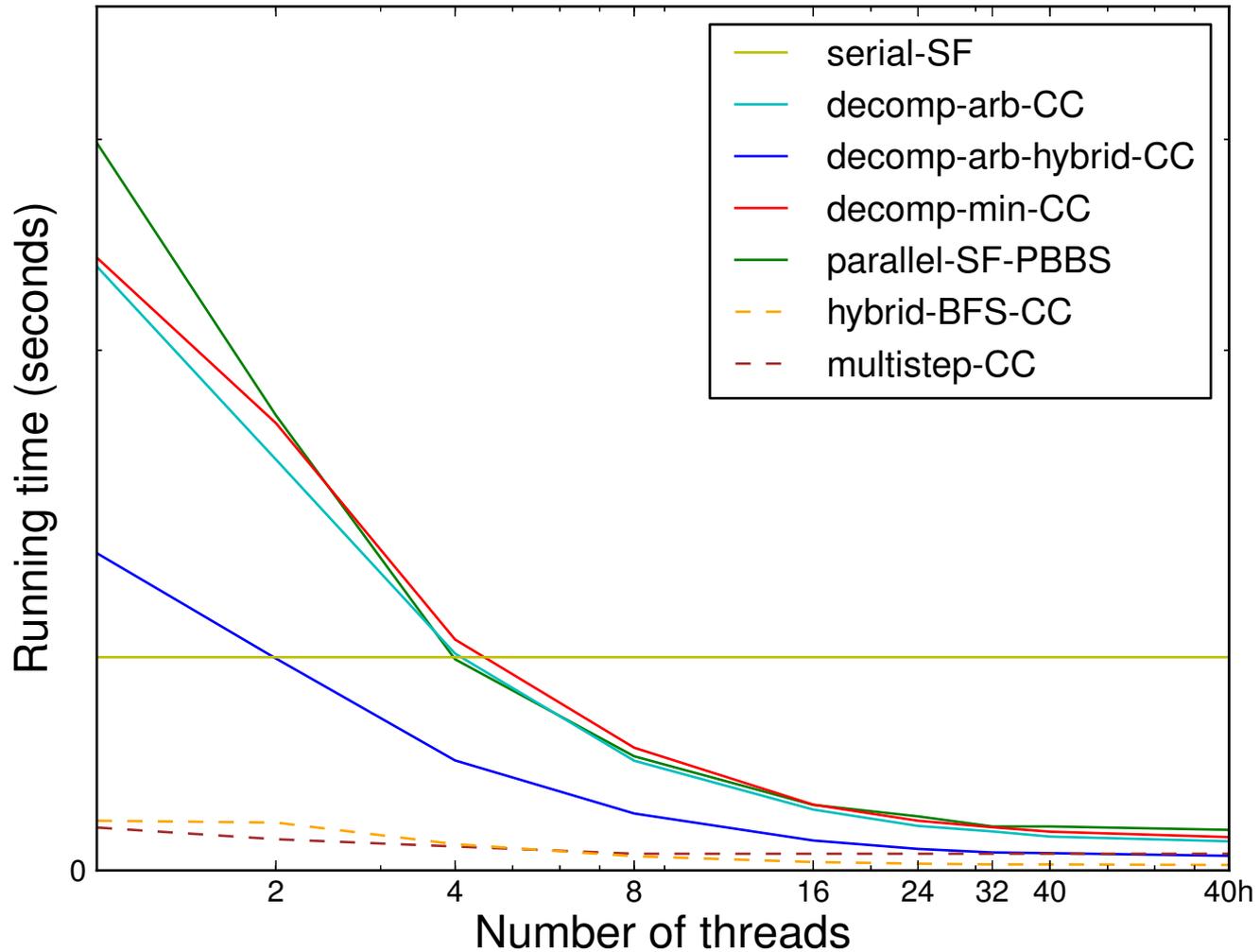
- Simple and practical linear-work, polylog-span connectivity algorithm
  - Can be easily modified to compute spanning forest
- As far as we know, first to be both practical and have linear work and polylog span
- Implementations competitive with existing parallel implementations
- Future direction: Can similar ideas give us a practical linear-work parallel algorithm for minimum spanning forest?

# Extra Slides

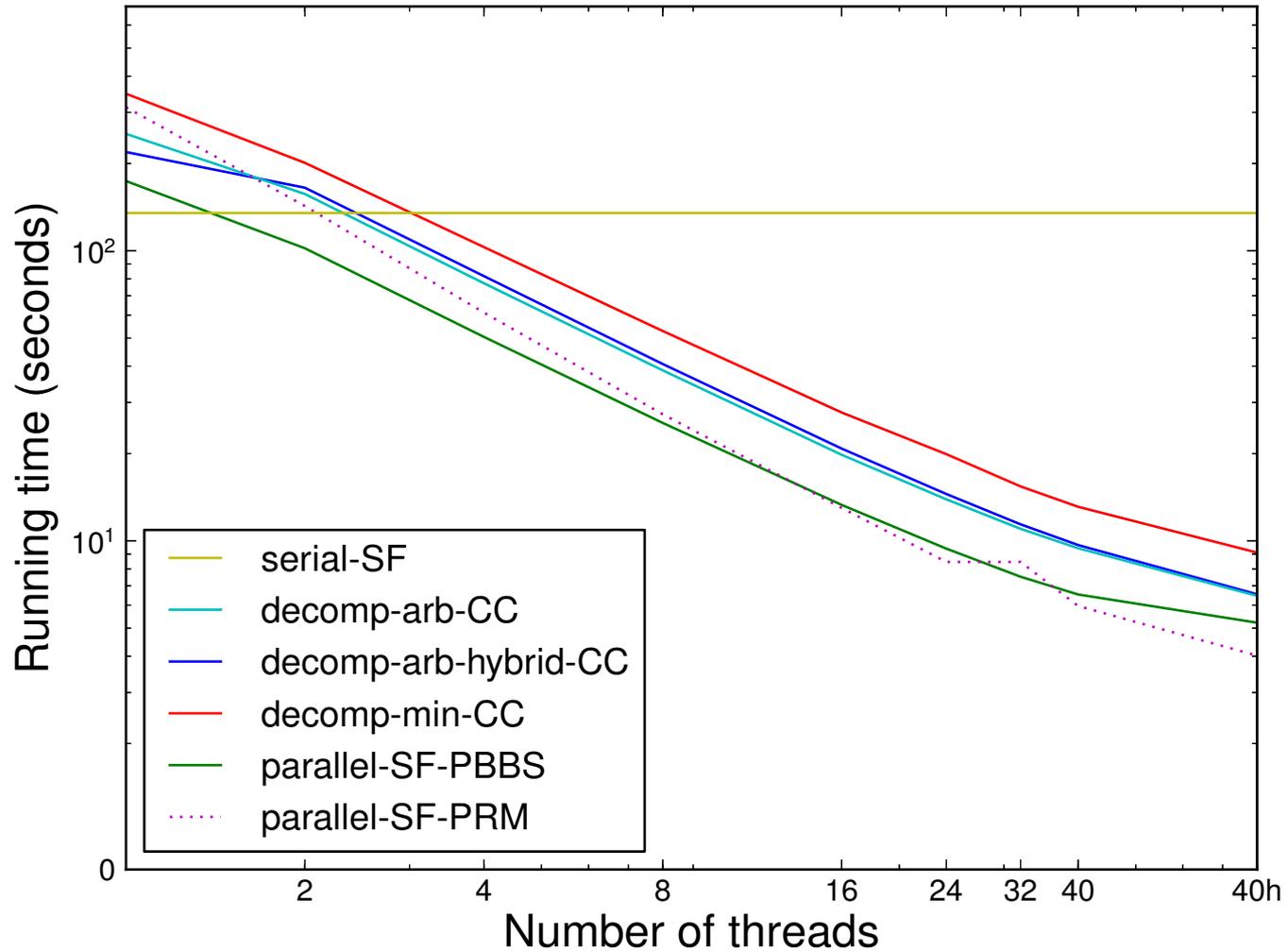
# 3D grid graph



# com-Orkut graph

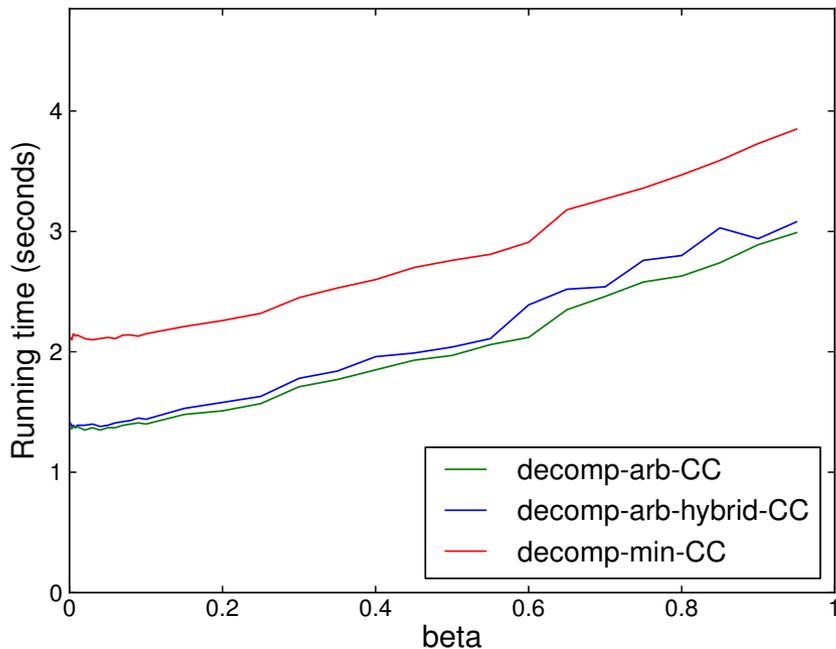


# Line graph

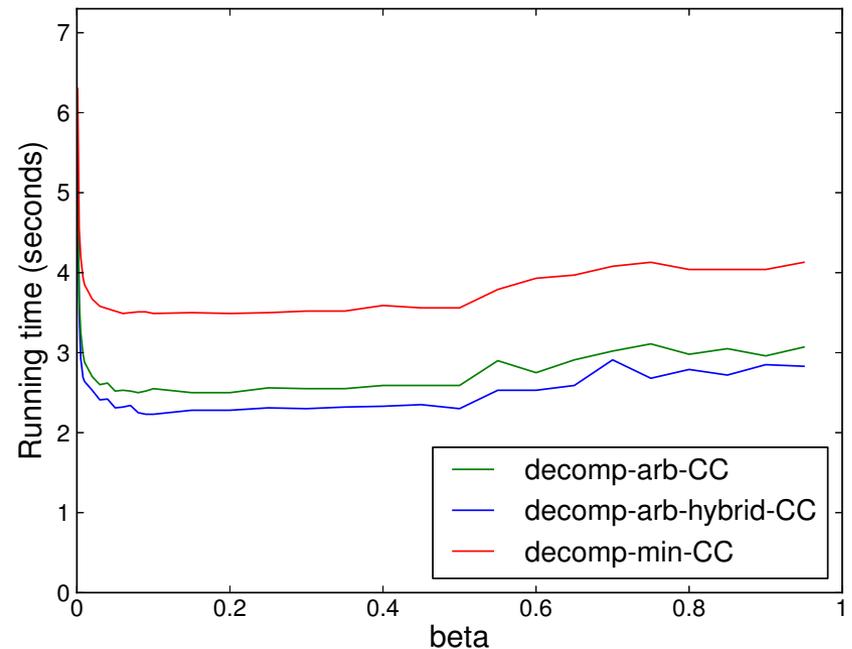


# Running time vs $\beta$

3D-grid graph



rMat graph



- Running time is similar across wide range of  $\beta$